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IN THE UNITED STATES PATENT AND TRADEMARK OFFICE

Patent Application

Applicant(s): Drissi et al.  
Docket No.: YOR920000401US1  
Serial No.: 09/713,342  
Filing Date: November 14, 2000  
Group: 2121  
Examiner: Wilbert L. Starks

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Signature: Vin Maurig Date: February 17, 2005

Title: Method and Apparatus for Generating a Data Classification  
Model Using an Adaptive Learning Algorithm

TRANSMITTAL OF CORRECTED APPEAL BRIEF

Mail Stop Appeal Brief - Patents  
Commissioner of Patents  
P.O. Box 1450  
Alexandria, VA 22313-1450

Sir:

Submitted herewith are the following documents relating to the above-identified patent application:

1. Appeal Brief; and
2. Copy of Notice of Appeal, filed on September 14, 2004, with copy of stamped return postcard indicating receipt of Notice by PTO on September 17, 2004.

There is no additional fee due in conjunction with this submission. In the event of non-payment or improper payment of a required fee, the Commissioner is authorized to charge or to credit **IBM Corporation's Deposit Account No. 50-0510** as required to correct the error. Duplicate copies of this letter and two copies of the Appeal Brief are enclosed.

Respectfully,

Kevin M. Mason

Date: February 17, 2005

Kevin M. Mason  
Attorney for Applicant(s)  
Reg. No. 36,597  
Ryan, Mason & Lewis, LLP  
1300 Post Road, Suite 205  
Fairfield, CT 06824  
(203) 255-6560



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5 Applicant(s): Drissi et al.  
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Signature: *Gene Maurin* Date: February 17, 2005

Title: Method and Apparatus for Generating a Data Classification Model Using an Adaptive Learning Algorithm

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CORRECTED APPEAL BRIEF

Mail Stop Appeal Brief - Patents  
20 Commissioner for Patents  
P.O. Box 1450  
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Sir:

25

Applicants hereby appeal the final rejection dated June 22, 2004, of claims 1 through 23 of the above-identified patent application.

REAL PARTY IN INTEREST

30

The present application is assigned to International Business Machines Corporation, as evidenced by an assignment recorded on November 14, 2000 in the United States Patent and Trademark Office at Reel 011307, Frame 0774. The assignee, International Business Machines Corporation, is the real party in interest.

35

RELATED APPEALS AND INTERFERENCES

There are no related appeals or interferences.

STATUS OF CLAIMS

Claims 1 through 23 are pending in the above-identified patent  
40 application. Claims 1-23 remain rejected as being directed to non-statutory subject

matter. Claims 8, 9, 21, and 23 remain rejected under 35 U.S.C. § 102(b) as being anticipated by McAulay, A.D. and Oh, J.C., Improved Learning in Genetic Rule-Based Classifier Systems, Systems, Man and Cybernetics, 1991; Decision Aiding for Complex Systems, Conference Proceedings, 1991 IEEE International Conference, October 13-16, 1991, Pages 1393-1398, Vol. 2 (hereinafter McAulay), and claims 1-23 remain rejected under 35 U.S.C. §103(a) as being unpatentable over McAulay et al. in view of Lewis, David D., An Evaluation of Phrasal and Clustered Representations on a Text Categorization Task, Proceedings of the Fifteenth Annual International ACM SIGIR Conference on Research and Development in Information Retrieval, June 1992, pages 37-50 (hereinafter Lewis).

#### STATUS OF AMENDMENTS

There have been no amendments filed subsequent to the final rejection.

#### SUMMARY OF CLAIMED SUBJECT MATTER

The present invention is directed to a data classification method and apparatus for labeling unknown objects. The disclosed data classification system employs a learning algorithm that adapts through experience. The present invention classifies objects in domain datasets using data classification models having a corresponding bias and evaluates the performance of the data classification. The performance values for each domain dataset and corresponding model bias are processed to identify or modify one or more rules of experience. (Page 9, line 4, to page 10, line 3.) The rules of experience are subsequently used to generate a model for data classification. Each rule of experience specifies one or more characteristics for a domain dataset and a corresponding bias that should be utilized for a data classification model if the rule is satisfied. (Page 10, lines 4-24.) The present invention dynamically modifies the assumptions (bias) of the learning algorithm to improve the assumptions embodied in the generated models and thereby improve the quality of the data classification and regression systems that employ such models. The disclosed self-adaptive learning process will become increasingly more accurate as the rules of experience are accumulated over time. (Page 10, line 25, to page 11, line 18.)

STATEMENT OF GROUNDS OF REJECTION TO BE REVIEWED ON APPEAL

Claims 1-23 are rejected as being directed to non-statutory subject matter; claims 8, 9, 21, and 23 are rejected under 35 U.S.C. § 102(b) as being anticipated by McAulay; and claims 1-23 are rejected under 35 U.S.C. §103(a) as being unpatentable  
 5 over McAulay et al. in view of Lewis.

ARGUMENTSection 101 Rejections

Claims 1-23 were rejected as being directed to non-statutory subject  
 10 matter. In particular, the Examiner asserts that claims 1, 8, 13, 16, and 21-23 are not claimed to be practiced on a computer and that it is clear that these claims are not limited to practice in the technological arts. The Examiner further asserts that none of the claims are limited to practical applications in the technological arts, that Applicants fail to define a useful, concrete and tangible result, and do not specify the associated practical  
 15 application with the appropriate level of specificity. The Examiner also finds that the Applicants manipulated a set of abstract "input data" to solve mathematical problems in the abstract and that the result of such manipulations is not statutory. Regarding the "system" and "computer readable medium" recitals in claims 16-23, the Examiner asserts that the invention is still found to be non-statutory.

20 Under Section 101, "any new and useful process, machine, manufacture, or composition of matter" is patentable. 35 U.S.C. §101. It is recognized, however, that despite the broad scope of section 101, "laws of nature, physical phenomena and abstract ideas" cannot be patented. *Diamond v. Chakrabarty*, 447 U.S. 303, 309, 206 U.S.P.Q. (BNA) 193, 197 (1980).

25 The Examiner asserts that Claims 1-23 are not claimed to be practiced on a computer and that it is clear that these claims are not limited to practice in the technological arts. To the contrary, however, each of the independent claims are expressly directed to a practical method of (or system for) "classifying data." For example, the method can be used to classify real numerical vectors. Thus, each of these  
 30 claims are clearly tied to a practical application. A process that is limited to a practical application of an abstract idea or mathematical algorithm in the technological arts is

patentable. *See* Examination Guidelines for Computer-Related Inventions, Section IV. B. 2. b. (ii).

In any event, the analysis does not stop there. The Supreme Court has stated that the "**[t]ransformation** and reduction of an article 'to a different state or thing' is the clue to patentability of a process claim." *Gottshalk v. Benson*, 409 U.S. 63, 70, 175 U.S.P.Q. (BNA) 676 (1972). In other words, claims that require some kind of transformation of subject matter, which has been held to include intangible subject matter, such as data or signals that are representative of or constitute physical activity or objects, have been held to comply with Section 101. *See, for example, In re Warmerdam*, 31 U.S.P.Q.2d (BNA) 1754, 1759 n.5 (Fed. Cir. 1994) or *In re Schrader*, 22 F.3d 290, 295, 30 U.S.P.Q.2d (BNA) 1455, 1459 n.12 (Fed. Cir. 1994).

Each independent claim includes at least one transformation. For example, independent claims 1, 16 and 22 **modify** the bias of one or more data classification models, based on a performance evaluation. Thus, a modified data classification model is provided. Claims 8, 21 and 23 **classify** objects and **select** a data classification model for classifying a domain dataset by comparing characteristics of the domain dataset to rules. Thus, an object classification is provided. Finally, claim 13 processes performance values for each combination of domain dataset and said bias to **adjust** one or more rules for subsequent data classification. Thus, adjusted rules are provided.

Applicants submit that each of the claims 1-23 are in full compliance with 35 U.S.C. §101, and accordingly, respectfully request that the rejection under 35 U.S.C. §101 be withdrawn.

#### Independent Claims 1, 8, 13, 16 and 21-23

Independent claims 8, 9, 21, and 23 are rejected under 35 U.S.C. § 102(b) as being anticipated by McAulay and independent claims 1, 8, 13, 16, and 21-23 are rejected under 35 U.S.C. §103(a) as being unpatentable over McAulay et al. in view of Lewis.

Regarding claim 1, the Examiner acknowledges that McAulay does not disclose selecting at least one of said one or more data classification models based on a meta-feature that characterizes said domain data set, but asserts that Lewis does show a

classifier using meta-features. Regarding claims 8, 21, and 23, the Examiner asserts that McAulay teaches selecting a data classification model for classifying a domain dataset by comparing characteristics of said domain dataset to said rules (FIG. 1: lines 4-5).

Regarding claim 1, Applicants note that Lewis teaches that “most current  
5 indexing languages represent documents as tuples or vectors of numeric or binary values, with *each value corresponding to an indexing term.*” (Page 38, Section 2.) Lewis then teaches that, “for clarity, we therefore call the features of indexing terms metafeatures.” (Page 38, Section 2.2). *Metafeatures in Lewis are therefore features of indexing terms* (the individual values representing a document) and not domain datasets. More  
10 importantly, Lewis does not disclose selecting data classification models based on a *meta-feature that characterizes a domain data set.* In addition, since Lewis only discloses the use of one algorithm (the genetic algorithm), there is no *selection of classification models.* Independent claims 1, 16, and 22 require classifying objects in a domain dataset using one or more data classification models, each of said one or more  
15 data classification models having a bias; selecting at least one of said one or more data classification models based on a meta-feature that characterizes said domain data set; evaluating the performance of said classifying step; and modifying said bias based on said performance evaluation. Independent claim 13 requires applying an adaptive learning algorithm to said domain dataset to select a data classification model based on a  
20 meta-feature that characterizes said domain data set, said data classification model having a bias; classifying objects in said domain dataset using said selected data classification model; evaluating the performance of said classifying step; maintaining an indication of said performance of said model for said domain dataset; repeating said applying, classifying and evaluating steps for a plurality of said domain datasets; and processing  
25 said performance values for each combination of said domain datasets and said bias to adjust one or more rules for subsequent data classification, each of said rules specifying one or more characteristics of said domain datasets and a corresponding bias that should be utilized in one of said data classification models. Independent claim 8, 21, and 23 require classifying objects in a plurality of domain datasets using one of a number of data  
30 classification models, each of said data classification models having a corresponding bias; evaluating the performance of each of said domain dataset classifications;

maintaining a performance value for each combination of said domain datasets and said bias; processing said performance values for each combination of said domain datasets and said bias to generate one or more rules, each of said rules specifying one or more characteristics of said domain datasets and a corresponding bias that should be utilized in one of said data classification models; and selecting a data classification model for classifying a domain dataset by comparing characteristics of said domain dataset to said rules.

Thus, McAulay et al. or Lewis, alone or in combination, do not disclose or suggest classifying objects in a domain dataset using one or more data classification models, each of said one or more data classification models having a bias; selecting at least one of said one or more data classification models based on a meta-feature that characterizes said domain data set; evaluating the performance of said classifying step; and modifying said bias based on said performance evaluation, as required by independent claims 1, 16, and 22, do not disclose or suggest applying an adaptive learning algorithm to said domain dataset to select a data classification model based on a meta-feature that characterizes said domain data set, said data classification model having a bias; classifying objects in said domain dataset using said selected data classification model; evaluating the performance of said classifying step; maintaining an indication of said performance of said model for said domain dataset; repeating said applying, classifying and evaluating steps for a plurality of said domain datasets; and processing said performance values for each combination of said domain datasets and said bias to adjust one or more rules for subsequent data classification, each of said rules specifying one or more characteristics of said domain datasets and a corresponding bias that should be utilized in one of said data classification models, as required by independent claim 13, and do not disclose or suggest classifying objects in a plurality of domain datasets using one of a number of data classification models, each of said data classification models having a corresponding bias; evaluating the performance of each of said domain dataset classifications; maintaining a performance value for each combination of said domain datasets and said bias; processing said performance values for each combination of said domain datasets and said bias to generate one or more rules, each of said rules specifying one or more characteristics of said domain datasets and a corresponding bias that should

be utilized in one of said data classification models; and selecting a data classification model for classifying a domain dataset by comparing characteristics of said domain dataset to said rules, as required by independent claims 8, 21, and 23.

#### Claims 3 and 18

5           Claims 3 and 18 were rejected under 35 U.S.C. §103(a) as being unpatentable over McAulay et al. in view of Lewis. The Examiner asserts that the limitation of claim 3 is taught by McAulay (FIG. 1: lines 4-5). Applicants note, however, that McAulay does not disclose or suggest generating one or more rules, *each of said rules specifying one or more characteristics of said domain datasets and a corresponding*  
10 *bias that should be utilized in one of said data classification models.*

Thus, McAulay et al. or Lewis, alone or in combination, do not disclose or suggest generating one or more rules, each of said rules specifying one or more characteristics of said domain datasets and a corresponding bias that should be utilized in one of said data classification models, as required by dependent claims 3 and 18.

#### Claims 4 and 19

15           Claims 4 and 19 were rejected under 35 U.S.C. §103(a) as being unpatentable over McAulay et al. in view of Lewis. The Examiner asserts that the limitation of claim 4 is taught by McAulay (Page 1393, third paragraph, first three lines of the paragraph). Applicants note, however, that McAulay does not disclose or suggest  
20 the step of selecting a data classification model for *classifying a domain dataset by comparing characteristics of said domain dataset to said rules.*

Thus, McAulay et al. or Lewis, alone or in combination, do not disclose or suggest the step of selecting a data classification model for classifying a domain dataset by comparing characteristics of said domain dataset to said rules, as required by  
25 dependent claims 4 and 19.

#### Conclusion

The rejections of the cited claims under §102 and §103 in view of McAulay et al. or Lewis, alone or in any combination, are therefore believed to be  
30 improper and should be withdrawn. The remaining rejected dependent claims are



believed allowable for at least the reasons identified above with respect to the independent claims.

The attention of the Examiner and the Appeal Board to this matter is appreciated.

5

Respectfully,



Date: February 17, 2005

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Kevin M. Mason  
Attorney for Applicant(s)  
Reg. No. 36,597  
Ryan, Mason & Lewis, LLP  
1300 Post Road, Suite 205  
Fairfield, CT 06824  
(203) 255-6560

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APPENDIX

1. A method for classifying data, comprising the steps of:  
classifying objects in a domain dataset using one or more data  
5 classification models, each of said one or more data classification models having a bias;  
selecting at least one of said one or more data classification models based  
on a meta-feature that characterizes said domain data set;  
evaluating the performance of said classifying step; and  
modifying said bias based on said performance evaluation.

10 2. The method of claim 1, wherein said steps of classifying and evaluating  
are performed for a plurality of said domain datasets and wherein said method further  
comprising the steps of recording a performance value for each combination of said  
domain datasets and said bias.

15 3. The method of claim 2, further comprising the step of processing said  
recorded performance values for each combination of said domain datasets and said bias  
to generate one or more rules, each of said rules specifying one or more characteristics of  
said domain datasets and a corresponding bias that should be utilized in one of said data  
20 classification models.

4. The method of claim 3, further comprising the step of selecting a data  
classification model for classifying a domain dataset by comparing characteristics of said  
domain dataset to said rules.

25 5. The method of claim 1, wherein said domain dataset is represented using a  
set of meta-features.

6. The method of claim 5, wherein said meta-features includes a concept  
30 variation meta-feature.

7. The method of claim 5, wherein said meta-features includes an average weighted distance meta-feature that measures the density of the distribution of said at least one domain dataset.

5 8. A method for classifying data, comprising the steps of:  
classifying objects in a plurality of domain datasets using one of a number  
of data classification models, each of said data classification models having a  
corresponding bias;  
evaluating the performance of each of said domain dataset classifications;  
10 maintaining a performance value for each combination of said domain  
datasets and said bias;  
processing said performance values for each combination of said domain  
datasets and said bias to generate one or more rules, each of said rules specifying one or  
more characteristics of said domain datasets and a corresponding bias that should be  
15 utilized in one of said data classification models; and  
selecting a data classification model for classifying a domain dataset by  
comparing characteristics of said domain dataset to said rules.

9. The method of claim 8, further comprising the step of modifying at least  
20 one of said biases based on said performance evaluation.

10. The method of claim 8, wherein said domain dataset is represented using a  
set of meta-features.

25 11. The method of claim 10, wherein said meta-features includes a concept  
variation meta-feature.

12. The method of claim 10, wherein said meta-features includes an average  
weighted distance meta-feature that measures the density of the distribution of said at  
30 least one domain dataset.

13. A method for classifying data in a domain dataset, comprising:  
 applying an adaptive learning algorithm to said domain dataset to select a  
 data classification model based on a meta-feature that characterizes said domain data set,  
 said data classification model having a bias;

5 classifying objects in said domain dataset using said selected data  
 classification model;

evaluating the performance of said classifying step;

maintaining an indication of said performance of said model for said  
 domain dataset;

10 repeating said applying, classifying and evaluating steps for a plurality of  
 said domain datasets; and

processing said performance values for each combination of said domain  
 datasets and said bias to adjust one or more rules for subsequent data classification, each  
 of said rules specifying one or more characteristics of said domain datasets and a  
 15 corresponding bias that should be utilized in one of said data classification models.

14. The method of claim 13, further comprising the step of selecting a data  
 classification model for classifying a domain dataset by comparing characteristics of said  
 domain dataset to said rules.

20 15. The method of claim 13, further comprising the step of modifying at least  
 one of said biases based on said performance evaluation.

16. A system for classifying data, comprising:  
 25 a memory that stores computer-readable code; and  
 a processor operatively coupled to said memory, said processor configured  
 to implement said computer-readable code, said computer-readable code configured to:

classify objects in a domain dataset using a one or more data classification  
 models, each of said one or more data classification models having a bias;

30 selecting at least one of said one or more data classification models based  
 on a meta-feature that characterizes said domain data set;

evaluate the performance of said classifying step; and  
modify said bias based on said performance evaluation.

17. The system of claim 16, wherein said processor is further configured to  
5 classify said objects and evaluate said performance for a plurality of said domain datasets  
and wherein said processor records a performance value for each combination of said  
domain datasets and said bias.

18. The system of claim 17, wherein said processor is further configured to  
10 process said recorded performance values for each combination of said domain datasets  
and said bias to generate one or more rules, each of said rules specifying one or more  
characteristics of said domain datasets and a corresponding bias that should be utilized in  
one of said data classification models.

15 19. The system of claim 18, wherein said processor is further configured to  
select a data classification model for classifying a domain dataset by comparing  
characteristics of said domain dataset to said rules.

20. The system of claim 16, wherein said domain dataset is represented using  
20 a set of meta-features.

21. A system for classifying data, comprising:  
a memory that stores computer-readable code; and  
a processor operatively coupled to said memory, said processor configured  
25 to implement said computer-readable code, said computer-readable code configured to:  
classify objects in a plurality of domain datasets using one of a number of  
data classification models, each of said data classification models having a corresponding  
bias;  
evaluate the performance of each of said domain dataset classifications;  
30 maintaining a performance value for each combination of said domain  
datasets and said bias;

process said performance values for each combination of said domain datasets and said bias to generate one or more rules, each of said rules specifying one or more characteristics of said domain datasets and a corresponding bias that should be utilized in one of said data classification models; and

5                   select a data classification model for classifying a domain dataset by comparing characteristics of said domain dataset to said rules.

22.               An article of manufacture for classifying data, comprising:

                  a computer readable medium having computer readable code means embodied thereon, said computer readable program code means comprising:

10                   a step to classify objects in a domain dataset using a one or more data classification models, each of said one or more data classification models having a bias;

                  selecting at least one of said one or more data classification models based on a meta-feature that characterizes said domain data set;

15                   a step to evaluate the performance of said classifying step; and

                  a step to modify said bias based on said performance evaluation.

23.               An article of manufacture for classifying data, comprising:

                  a computer readable medium having computer readable code means embodied thereon, said computer readable program code means comprising:

20                   a step to classify objects in a plurality of domain datasets using one of a number of data classification models, each of said data classification models having a corresponding bias;

                  a step to evaluate the performance of each of said domain dataset classifications;

25                   a step to maintaining a performance value for each combination of said domain datasets and said bias;

                  a step to process said performance values for each combination of said domain datasets and said bias to generate one or more rules, each of said rules specifying one or more characteristics of said domain datasets and a corresponding bias that should be utilized in one of said data classification models; and

a step to select a data classification model for classifying a domain dataset by comparing characteristics of said domain dataset to said rules.



 **COPY**



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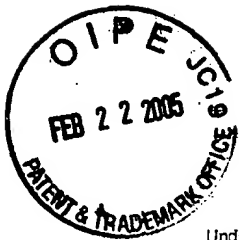
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September 14, 2004  
Serial No.: 09/713,342  
YOR920000401US1  
1500-144 (KMM)





PTO/SB/31 (02-01)  
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**NOTICE OF APPEAL FROM THE EXAMINER TO THE  
BOARD OF PATENT APPEALS AND INTERFERENCES**

Docket Number (Optional)

YOR920000401US1

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Signature Tina Maurice

Typed or printed name Tina Maurice

In re Application of

Drissi et al.

Application Number

09/712,638

Filed

November 14, 2000

For Method and Apparatus for Generating a Data Classification Model Using an Adaptive Learning Algorithm

Group Art Unit

2121

Examiner

Wilburt L. Starks

Applicant hereby **appeals** to the Board of Patent Appeals and Interferences from the last decision of the examiner.

The fee for this Notice of Appeal is (37 CFR 1.17(b))

\$ 330.00

- ☐ Applicant claims small entity status. See 37 CFR 1.27. Therefore, the fee shown above is reduced by half, and the resulting fee is: \$\_\_\_\_\_
- ☐ A check in the amount of the fee is enclosed.
- ☐ Payment by credit card. Form PTO-2038 is attached.
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- ☒ The Commissioner is hereby authorized to charge any fees which may be required, or credit any overpayment to Deposit Account No. 50-0510. I have enclosed a duplicate copy of this sheet.
- ☒ A petition for an extension of time under 37 CFR 1.136(a) (PTO/SB/22) is enclosed.

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I am the

- ☐ applicant/inventor.
- ☐ assignee of record of the entire interest.  
See 37 CFR 3.71. Statement under 37 CFR 3.73(b) is enclosed. (Form PTO/SB/96)
- ☒ attorney or agent of record.
- ☐ attorney or agent acting under 37 CFR 1.34(a).  
Registration number if acting under 37 CFR 1.34(a): \_\_\_\_\_

Kevin M. Mason  
Signature

Kevin M. Mason  
Typed or printed name

September 14, 2004  
Date

NOTE: Signatures of all the inventors or assignees of record of the entire interest or their representative(s) are required. Submit multiple forms if more than one signature is required, see below\*.

☐ \*Total of 1 forms are submitted.

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